

# Some definitions

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September 26, 2017

In this note I will write down some of the definitions that are often used in the course.

**Definition.** Courant condition

The Courant condition (also called Courant-Friedrichs-Lewy (CFL) condition) is a condition that must (often) hold for the finite difference scheme to be stable.

For the two dimensional wave equation

$$\frac{\partial^2 u}{\partial t^2} = q(x, y) \left( \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} \right)$$

it can be deduced that the Courant condition is

$$c^2 \left( \frac{dt^2}{dx^2} + \frac{dt^2}{dy^2} \right) \leq 1,$$

where  $c = \sqrt{\max q(x)}$ . This gives the stability condition

$$dt \leq \frac{1}{c} \frac{1}{\sqrt{\frac{1}{dx^2} + \frac{1}{dy^2}}}$$

on  $dt$ . Usually we let  $dt$  be a little smaller than this.

**Definition.** Neumann/Dirichlet boundary conditions

A von Neumann boundary condition states that the solution to the PDE should satisfy

$$\frac{\partial u}{\partial \mathbf{n}}(\mathbf{x}) = f(\mathbf{x}) \quad \forall \mathbf{x} \in \partial\Omega,$$

where  $\Omega$  is the domain where the function  $u$  is defined. Often the function  $f$  is equal to 0. The normal derivative is defined by

$$\frac{\partial y}{\partial \mathbf{n}}(\mathbf{x}) = \nabla y(\mathbf{x}) \cdot \mathbf{n}(\mathbf{x}),$$

where  $\mathbf{n}(\mathbf{x})$  denotes the outward pointing normal vector.

While the von Neumann condition gives restrictions on the derivative of the solution at the boundary, a Dirichlet boundary condition says what the value of the solution should be at the boundary:

$$u(\mathbf{x}) = f(\mathbf{x}) \quad \forall \mathbf{x} \in \partial\Omega.$$

**Definition.** Convergence rate

For many numerical schemes we have the relation

$$E = Ch^r$$

where  $E$  is the error,  $h$  is some parameter proportional to the mesh size, and  $C$  and  $r$  are constants. The convergence rate of the scheme is  $r$ . To find  $r$  we consider the error with two different  $h$  values, say  $(E_i, h_i)$  and  $(E_{i+1}, h_{i+1})$ . Since both errors should satisfy the relation above, we have

$$\begin{aligned} E_i &= Ch_i^r, \\ E_{i+1} &= Ch_{i+1}^r. \end{aligned}$$

Divide the two equations and take the logarithm,

$$\frac{E_i}{E_{i+1}} = \left( \frac{h_i}{h_{i+1}} \right)^r,$$

$$\log(E_i/E_{i+1}) = r \log(h_i/h_{i+1}),$$

$$r = \frac{\log(E_i/E_{i+1})}{\log(h_i/h_{i+1})}.$$

The  $r$  above is often denoted  $r_i$ , and you should compute a series of  $r_i$ 's to see if the convergence rate is stabilizing when  $h$  is getting small.

Note that if  $h_i$  and  $h_{i+1}$  are very small you may get problems due to precision errors when the computer is dividing small numbers.